Taming the complexity of biochemical networks through model reduction and tropical geometry

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Polynomial kinetics (e.g. mass action law)

$$\frac{dx_i}{dt} = P_i(\boldsymbol{x}) = \sum_{j=1}^{M_i} k_{ij} \boldsymbol{x}^{\alpha_{ij}}, \ \boldsymbol{x}^{\alpha_{ij}} = x_1^{\alpha_{ij}^1} \dots x_n^{\alpha_{ij}^n}.$$

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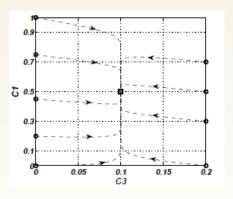
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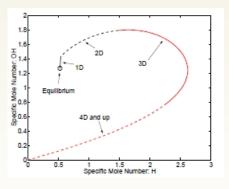
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- Which details of the model can be neglected? Problem: small terms can not always be neglected.
- Can one provide symbolic descriptions of the reduced model dynamics?

Chemical reaction networks Slow/fast systems







from Chiavazzo et al Comm.Comp.Phys. 2007 from Hung and Sheperd 12th Ann. Int. Detonation Symp. 2002 In fast/slow systems, fast variables relax quickly, then they are slaved: the system moves on a low dimensional invariant manifold describing the reduced model. The reduced model can change with time.

Chemical reaction networks > Tikhonov theorem



$$\frac{dx}{dt} = \frac{1}{\varepsilon} f(x, y)$$
(1: fast)

$$\frac{dy}{dt} = g(x, y)(2: slow)$$

Tikhonov: If for any y the dynamics (1) has a hyperbolic point attractor, then after a fast transition the system evolves according to:

$$\frac{dy}{dt} = g(x, y)$$

and

f(x,y)=0 the fast variables are slaved by slow ones





Several steps:

 Determine the fast/slow decomposition : partially solved by CSP (Lam and Goussis 1994), implemented in COPASI.



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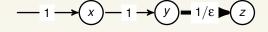


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- Pool species : use conservation laws (Gorban, Radulescu, Zinovyev 2010).

Chemical reaction networks Quasi-steady state approximation



Fast, low concentration radicals



$$\frac{dx}{dt} = 1 - x$$

$$\frac{dy}{dt} = x - \frac{1}{\varepsilon}y$$

$$\frac{dz}{dt} = \frac{1}{\varepsilon}y - z$$

$$y = \varepsilon x$$

$$y=\varepsilon x$$

y is fast, x, z are slow.

Chemical reaction networks Ouasi-steady state approximation





$$\frac{dx}{dt} = 1 - x$$

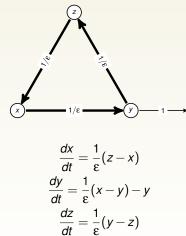
$$\frac{dz}{dt} = x - z$$

Reduced model: prune fast species, replace fast sub-system by elementary modes (reaction pooling)

Chemical reaction networks • Quasi-equilibrium approximation



Fast cycles (all the species in the cycle are fast)



Conservation laws of the fast subsystem that are not conserved by the full system are slow.

Chemical reaction networks • Quasi-equilibrium approximation



$$(x+y+z)$$
 1/3 \rightarrow

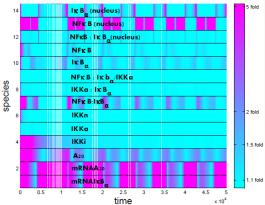
$$\frac{d(x+y+z)}{dt} = -1/3(x+y+z)$$

Reduced model : prune fast reactions, pool species.

Chemical reaction networks > Slowness index : detecting fast and slow variables



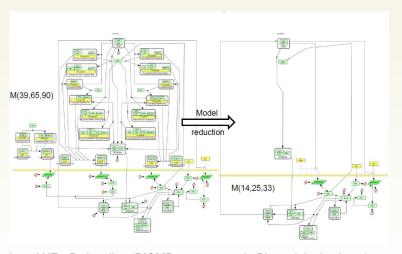
Let x^* be the solution of f(x, y(t)) = 0 where y(t) are simulated trajectories. The slowness index for x is $I(t) = |\log(x(t)/x^*(t))|$.



Slowness index calculated for the species in the NFκB model BIOMD000000226.

Chemical reaction networks > An example





Reduction of NF- κ B signaling (BIOMD000000227 in Biomodels database)

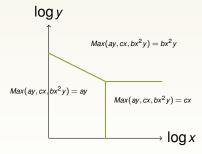
Tropical methods → Tropical curves and varieties



Consider a polynomial in two variables, say $ay + cx + bx^2y$.

 $\text{Max-plus algebra} + \rightarrow \textit{Max}, \times \rightarrow +.$

A tropical curve is the set of points where the corresponding max-plus polynomial Max(log(a) + y, log(c) + x, log(b) + 2log(x) + log(y)) is not smooth.



A tropical curve is an algebraic curve defined over the max-plus semifield.

Tropical methods Newton-Puiseux series



Solve $\varepsilon y - x + \varepsilon^2 x^2 y = 0$, ε small positive parameter



Tropical methods Newton-Puiseux series



- Solve $\varepsilon y x + \varepsilon^2 x^2 y = 0$, ε small positive parameter
- Kapranov theorem: there are solutions $x = x_1 \varepsilon^{a_1} + x_2 \varepsilon^{a_2} + \ldots$, $y = y_1 \varepsilon^{b_1} + y_2 \varepsilon^{b_2} + \ldots$, where $a_1, a_2, \ldots, b_1, b_2, \ldots$ are increasing rational numbers with a common denominator.

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- » must have, at lowest order $y_1 \varepsilon^{1+b_1} x_1 \varepsilon^{a_1} + x_1^2 y_1 \varepsilon^{2a_1+b_1} = 0$

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- to have real solutions need $a_1 = min(1 + b_1, 2a_1 + b_1)$ Tropical equilibration: the positive and negative dominant terms have the same orders.

Tropical methods



Newton-Puiseux series

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Tropical methods > Newton-Puiseux series



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$$y - x - \varepsilon x^4 = 0$$
$$x - y + \varepsilon y^2 = 0$$
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- \rightarrow Only (0,0) and (-1/2,-1) can be lifted to Newton-Puiseux series.

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- By adding the equations, get $y^2 x^4 = 0$ leading to $a_1 = a_2/2$. Only two solutions left, (0,0) and (-1/2,-1).

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- By adding the equations, get $y^2 x^4 = 0$ leading to $a_1 = a_2/2$. Only two solutions left, (0,0) and (-1/2,-1).
- > In general, have to compute a tropical basis : hard.



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Define $x_i = \varepsilon^{a_i} \bar{x}_i$ where a_i are unknown.

$$\frac{d\bar{x}_i}{dt} = \sum_{j=1}^{M_i} \varepsilon^{\mu_{ij}} \bar{k}_{ij} \bar{\mathbf{x}}^{\alpha^{ij}},$$

$$\mu_{ij} = \gamma_{ij} + \sum_{l=1}^{n} \alpha_l^{ij} a_l - a_i.$$



At least two terms, of opposite signs have the same degree in ϵ .

Find a pair (j,j'), $j \neq j'$ such that

- $\mathsf{i)}\ \mu_{\mathsf{i}\mathsf{j}} = \mu_{\mathsf{i}\mathsf{j}'},$
- ii) $\mu_{ij} \leq \mu_{il}$ for all $l \neq j, j'$,
- iii) $k_{ij}k_{ij'}<0$.

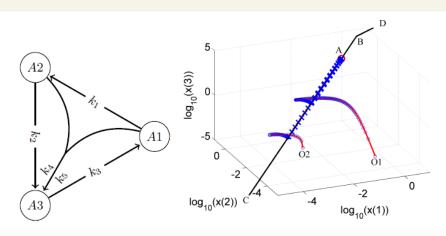
The minimum degree is attained at least twice and the minimal degree terms have opposite signs.

$$rac{dar{\mathbf{x}}_i}{dt} = \mathbf{\epsilon}^{\mu_i}(|ar{\mathbf{k}}_{ij}|ar{\mathbf{x}}^{oldsymbol{lpha}^{ij}} - |ar{\mathbf{k}}_{ij'}|ar{\mathbf{x}}^{oldsymbol{lpha}^{ij'}})$$

The tropically truncated system is generically binomial (toric system).

Nonlinear cycle example

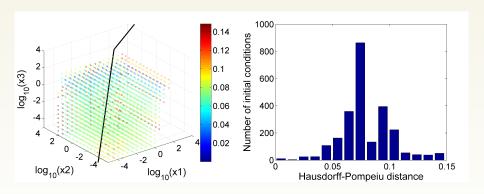




$$\begin{array}{l} \gamma_1 = 0, \gamma_2 = 1, \gamma_3 = \gamma_4 = 2, \gamma_5 = 3 \\ \min(\gamma_1 + a_1, \gamma_4 + a_1 + a_2) = \min(\gamma_3, \gamma_5) + a_3 = \min(\gamma_2 + a_2, \gamma_4 + a_1 + a_2) = \\ \min(\gamma_1 + a_1, \gamma_5 + a_3) \end{array}$$

Nonlinear cycle example





Michaelis-Menten enzymatic reaction



$$S+E \underset{k_{-1}}{\overset{k_1}{\rightleftharpoons}} ES \overset{k_2}{\Rightarrow} P+E,$$

Using the conserved quantities $e_0 = [E] + [ES]$, $s_0 = [S] + [ES] + [P]$, we get the system

$$x' = -k_1x(e_0 - y) + k_{-1}y, \quad y' = k_1x(e_0 - y) - (k_{-1} + k_2)y.$$

where x = [S] and y = [SE], and some constraints

$$0 \leq y \leq e_0, \, 0 \leq x+y \leq s_0, \, 0 \leq x.$$

Michaelis-Menten enzymatic reaction The tropical equilibration problem



After rescalings $x = \bar{x} \varepsilon^{a_1}$, $y = \bar{y} \varepsilon^{a_2}$, $k_1 = \bar{k}_1 \varepsilon^{\gamma_1}$, $k_{-1} = \bar{k}_{-1} \varepsilon^{\gamma_{-1}}$, $e_0 = \bar{e}_0 \varepsilon^{\gamma_e}$, $s_0 = \bar{s}_0 \varepsilon^{\gamma_s}$, we get

$$\begin{split} \bar{x}' &= -\bar{k}_1 \bar{e}_0 \epsilon^{\gamma_1 + \gamma_e} \bar{x} + \bar{k}_1 \epsilon^{\gamma_1 + a_2} \bar{x} \bar{y} + \bar{k}_{-1} \epsilon^{\gamma_{-1} + a_2 - a_1} \bar{y}, \\ \bar{y}' &= \bar{k}_1 \bar{e}_0 \epsilon^{\gamma_1 + \gamma_e + a_1 - a_2} \bar{x} - \bar{k}_1 \epsilon^{\gamma_1 + a_1} \bar{x} \bar{y} - (\bar{k}_{-1} \epsilon^{\gamma_{-1}} + \bar{k}_2 \epsilon^{\gamma_2}) \bar{y}. \end{split}$$

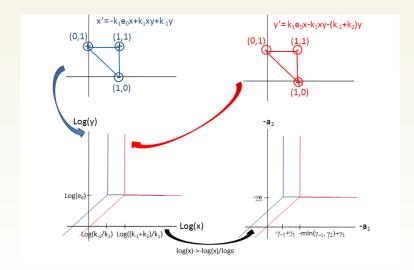
that leads to the equilibration equations

$$\gamma_1 + \gamma_e = \min(\gamma_1 + a_2, \gamma_{-1} + a_2 - a_1),$$

 $\gamma_1 + \gamma_e + a_1 - a_2 = \min(\gamma_1 + a_1, \min(\gamma_{-1}, \gamma_2)).$

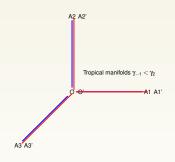
Michaelis-Menten enzymatic reaction → Newton Polygons and Tropical Manifolds





Michaelis-Menten enzymatic reaction Quasi-equilibrium approximation

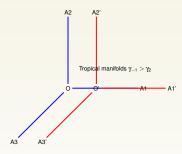




No	Condition	Truncated system	Regime
1	$a_1 < \gamma_{-1} - \gamma_1$ $a_2 = \gamma_e$	$x' = \varepsilon^{\gamma_1 + \gamma_e} (-\bar{k}_1 \bar{e}_0 \bar{x} + \bar{k}_1 \bar{x} \bar{y})$ $y' = \varepsilon^{\gamma_1 + a_1} (\bar{k}_1 \bar{e}_0 \bar{x} - \bar{k}_1 \bar{x} \bar{y})$	QE saturated
2	$a_1 > \gamma_{-1} - \gamma_1$ $a_2 = a_1 + \gamma_e + \gamma_1 - \gamma_{-1}$	$x' = \varepsilon^{\gamma_1 + \gamma_e} \left(-\overline{k}_1 \overline{e}_0 \overline{x} + \overline{k}_{-1} \overline{y} \right)$ $y' = -\varepsilon^{\gamma_{-1}} \left(\overline{k}_1 \overline{e}_0 \overline{x} - \overline{k}_{-1} \overline{y} \right)$	QE linear

Michaelis-Menten enzymatic reaction → Quasi-steady state approximation





No	Condition	Truncated system	Regime
1	$a_1 < \gamma_2 - \gamma_1$ $a_2 = \gamma_e$	$x' = \varepsilon^{\gamma_1 + \gamma_e} (-\bar{k}_1 \bar{e}_0 \bar{x} + \bar{k}_1 \bar{x} \bar{y})$ $y' = \varepsilon^{\gamma_1 + a_1} (\bar{k}_1 \bar{e}_0 \bar{x} - \bar{k}_1 \bar{x} \bar{y})$	QE
2	$\begin{vmatrix} \gamma_2 - \gamma_1 < a_1 < \gamma_{-1} - \gamma_1 \\ a_2 = \gamma_e \end{vmatrix}$	$x' = \varepsilon^{\gamma_1 + \gamma_e} (-\bar{k}_1 \bar{e}_0 \bar{x} + \bar{k}_1 \bar{x} \bar{y})$ $y' = -\varepsilon^{\gamma_2} \bar{k}_2 \bar{y}$	x QSS
3	$a_1 > \gamma_{-1} - \gamma_1$ $a_2 = a_1 + \gamma_e + \gamma_1 - \gamma_{-1}$	$x' = \varepsilon^{\gamma_1 + \gamma_e} (-\bar{k}_1 \bar{e}_0 \bar{x} + \bar{k}_{-1} \bar{y})$ $y' = -\varepsilon^{\gamma_2} \bar{k}_2 \bar{y}$	x QSS
4	$a_1 > \gamma_2 - \gamma_1$ $a_2 = a_1 + \gamma_e + \gamma_1 - \gamma_2$	$x' = -\varepsilon^{\gamma_1 + \gamma_{\bar{e}}} \bar{k}_1 \bar{e}_0 \bar{x}$ $y' = \varepsilon^{\gamma_2} (\bar{k}_1 \bar{e}_0 \bar{x} - \bar{k}_2 \bar{y})$	y QSS

Michaelis-Menten enzymatic reaction Tropical equilibrations and model reduction



TTS at quasi-equilibrium: pruning

$$\bar{x}' = \varepsilon^{\gamma_1 + \gamma_e} (-\bar{k}_1 \bar{e}_0 \bar{x} + \bar{k}_{-1} \bar{y})$$

$$\bar{y}' = \varepsilon^{\gamma_{-1}} (\bar{k}_1 \bar{e}_0 \bar{x} - \bar{k}_{-1} \bar{y})$$

New slow variable z = x + y: pooling

$$\bar{z}' = -\varepsilon^{\gamma_2 + a_2 - \gamma_s} \bar{k}_2 \bar{y}$$

Eliminate x, y, obtain the reduced model

$$z' = -k_2/(1 + k_{-1}/(k_1 e_0))z.$$

Results on Biomodels.net



436 curated models, 55 have purely polynomial kinetics.

Two methods:

Reified constraints

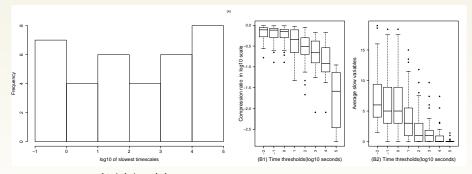
Found	# models	Variables (avg/min/max)	Time (avg/min/max)
yes	23	17.348/3/ 86	0.486/0.004/2.803
no	32	17.812/1/194	0.099/0.000/1.934

Newton polytope

ε	Total	Timed-	Models	Models	Average
value	models	out	without	with	running
	consid-	models	tropical	tropical	time (in
	ered		solu-	solu-	secs)
			tions	tions	
1/5	53	16	0	37	405.50
1/7	53	16	0	37	734.74
1/9	53	16	0	37	716.40
1/23	50	17	1	32	889.59

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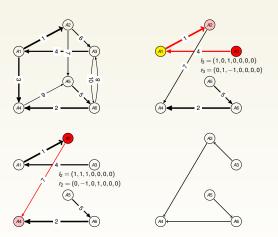


 $\mu_{\text{threshold}} = -\log(\theta/\tau)/\log(\epsilon), \, \tau \text{ time units.}$

Species with $\mu_i - a_i \ge \mu_{\text{threshold}}$ are slow.

At timescales of 1000s models have median numbers of 2 slow variables.

> Monomolecular case



Any monomolecular network with separated constants can be reduced to an acyclic, deterministic digraph (Gorban and Radulescu 2007). The symbolic dynamics can described by a deterministic finite state automaton. The number of states is at most the number of species.

> Nonlinear networks

> Main idea : use tropical equilibration branches (slow) as proxys for metastability.



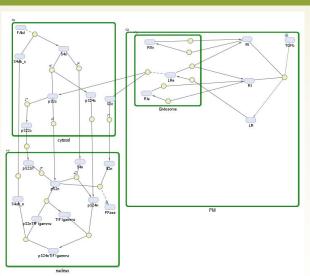
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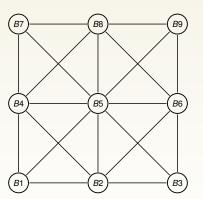


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- > We define minimal branches: minimal truncated system.
- Minimal branches are algebraically connected if they have a common overset.



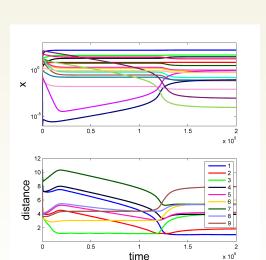


> Nonlinear networks



Minimal tropical branches and algebraic connections.

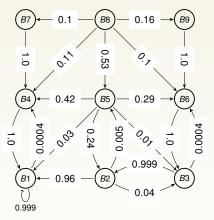
> Nonlinear networks



Numerically computed trajectories and distances to minimal branches.

> Nonlinear networks





Stochastic finite state automaton learned from trajectories with various initial conditions.

Conclusion



- Solving the tropical equilibration problem allows fast/slow decomposition and model reduction of biochemical reactions networks.
- We have two methods for solving the tropical equilibration problem, a first one by reformulating it as a constraint satisfaction problem and a second one based on the Newton polyhedron.
- Tropical methods can be also used to coarse grain the dynamics of a smooth biochemical reaction network to a discrete symbolic dynamics of a finite state automaton.

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